Extending Kolmogorov's axioms for collections of contexts

http://tph.tuwien.ac.at/~svozil/publ/2022-QIP22-pres.pdf based on arXiv:1903.10424

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Announcement: "Varieties of contextuality based on probability and structural nonembeddability"

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Challenge (spontaneous & unrelated to main topic of talk): Find a True-Implies-False (TIFS) gadget with "aperture" $\pi/2$

In 1965 Kochen & Specker introduced the "Specker bug" gadget graph in DOI 10.1007/978-3-0348-9259-9_19.

In their better known 1967 paper DOI 10.1512/iumj.1968.17.17004 they used the "Specker bug" gadget to construct a True-Implies-True (TITS) gadget.

They employed the latter in proofs of (i) "the Kochen & Specker theorem", as well as a (ii) configuration of quantum observables that cannot be separated by classical means (ie, two-valued states), indicating non-imbeddibility, a very strong form of "contextuality".

(Subsequently this "Specker bug" has been independently re-discovered a couple of times.)

Challenge (spontaneous & unrelated to main topic of talk): Find a True-Implies-False (TIFS) gadget with "aperture" $\pi/2$ cntd.

For all True-Implies-False gadgets known so far-eg, reviewed in

- Cabello, Portillo, Solís and KS, DOI 10.1103/PhysRevA.98.012106,
- Abbott, Calude and KS, DOI 10.1063/1.4931658,
- Ramanathan, Rosicka, Pironio, Karol, Michal, and Pawel Horodecki, DOI 10.22331/q-2020-08-14-308,

the TIFS "aperture" beween its end points is less than $\pi/2$.

(TITS with "aperture" $\pi/2$ exist plentiful "by serial composition" and are used in proofs of "the Kochen & Specker theorem".)

Challenge (spontaneous & unrelated to main topic of talk): Find a True-Implies-False (TIFS) gadget with "aperture" $\pi/2$ cntd.

Challenge of Mohammad H. Shekarriz and KS, "Noncontextual coloring of orthogonality hypergraphs", Journal of Mathematical Physics 63 (3), 032104 (2022) DOI 10.1063/5.0062801:

"Find a TIFS containing no TITS as subgadget whose "aperture" is exactly $\pi/2$.

Or, alternatively, prove that this is impossible."

ps: trivial TIFS like a single context—there should at least be two intertwining contexts, or TITS-TIFS combos, are excluded.

pps: Any proof that a faithful orthogonal representation of a TIFS with aperture $\pi/2$ might ultimately require geometric means. Maybe such a proof is elementary or hard, I have not intuition about it.

• Schrödinger, on p. 15 of "My View of the World", quoted the Vedantic analogy of a "many-faceted crystal which, while showing hundreds of little pictures of what is in reality a single existent object, does not really multiply that object. . . . A comparison used in Hinduism is of the many almost identical images which a many-faceted diamond makes of some one object such as the sun."

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- Another example from partition logic—aka automaton logik aka Wright's generalized urn model: the different "ball states" (aka partitions referring to collections of ball types) when "viewed by different color filters".
- This is about epistemology, since either the ontology "remains hidden" as it cannot be accessed by physical means in the presence of complementarity, or is restricted to merely one maximal observable aka context.

Informal notion of maximal observables aka contexts

A "view" or (used synonymously) "frame" or "context" will be in full generality and thus informally (glancing at heuristics from quantum mechanics and partition logic) characterized as some domain or set of observables or properties which are

- (i) *largest* or *maximal* in the sense that any extension yields redundancies,
- (ii) yet at the same time in the finest resolution in the sense that the respective observables or properties are "no composite" of "more elementary" ones,
- (iii) mutually exclusive in the sense that one property or observation excludes another, different property or observation, as well as
- (iv) contains only *simultaneously measurable*, *compatible* observables or properties.

Generalization of Kolmogorov's axioms to arbitrary event structures

Suppose two arbitrary contexts $\mathcal{C}_1 = \{e_1, \dots e_n\}$ and $\mathcal{C}_2 = \{f_1, \dots f_m\}$. The conditional probabilities $P(f_j|e_i)$, with $1 \leq j \leq m$ and $1 \leq i \leq n$, which alternatively can be considered as either measuring the Bayesian degree of reasonable expectation representing a state of knowledge or as quantification of a personal belief or the frequency of occurrence of " f_j given e_i ", can be arranged into a $(n \times m)$ -matrix whose entries are $P(f_j|e_i)$, that is,

$$[P(\mathcal{C}_{2}|\mathcal{C}_{1})] = [P(\{f_{1}, \dots f_{m}\}|\{e_{1}, \dots e_{n}\})]$$

$$\equiv \begin{bmatrix} P(f_{1}|e_{1}) & \cdots & P(f_{m}|e_{1}) \\ \cdots & \cdots & \cdots \\ P(f_{1}|e_{n}) & \cdots & P(f_{m}|e_{n}) \end{bmatrix}. \tag{1}$$

Generalization of Kolmogorov's axioms to arbitrary event structures cntd.

Assume that the conditional probabilities of the elements of the second context with respect to an arbitrary element $e_k \in \mathcal{C}_1$ of the first context \mathcal{C}_1 are non-negative, additive, and that, if this sum is extended over the entire second context \mathcal{C}_2 , it adds up to one:

$$P(f_i|e_k) + P(f_j|e_k) = P[(f_i \cup f_j)|e_k]$$

$$\sum_{f_i \in \mathcal{C}_2} P(f_i|e_k) = P\left[\left(\bigcup_{f_i \in \mathcal{C}_2} f_i\right)|e_k\right] = 1.$$
(2)

That is, the row sum taken within every single row of $[P(C_2|C_1)]$ adds up to one.

This generalises Kolmogorov's axioms as it allows cases in which both contexts do not coincide.

Examples: Quantum bistochasticity

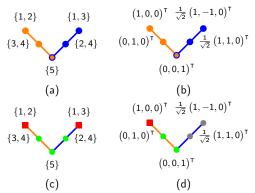
The multi-context quantum case has been studied in great detail with emphasis on motivating and deriving the Born rule from elementary foundations by Alexia Aufféves and Philippe Grangier, eg (among others),

- DOI 10.1038/srep43365
- DOI 10.1098/rsta.2017.0311

Please look at their papers!

Examples: Two intertwining three-atomic contexts—the "firefly" logic L_{12}

Hypergraph of the L_{12} "firefly" logic. (a) The associated (quasi)classical partition logic representation obtained through in inverse construction using all two-valued measures thereon; (b) a faithful orthogonal representation rendering a quantum *double*; (c) "classical" two-valued measure number 1; (d) a pure quantum state prepared as $(1,0,0)^{\mathsf{T}}$. A red square and gray and green circles indicate value assignments 1, $\frac{1}{2}$ and 0, respectively.





Examples: Classical probabilities on two intertwining three-atomic contexts—the "firefly" logic L_{12}

The L_{12} "firefly" logic labels the atoms (aka elementary propositions) obtained by an "inverse construction" using all five two-valued measures thereon. By design, it will be very similar to the earlier logic with four atoms. With the identifications $\mathbf{e}_1 \equiv \{1,2\}$, $\mathbf{e}_2 \equiv \{3,4\}$, $\mathbf{e}_3 = \mathbf{f}_3 \equiv \{5\}$, $\mathbf{f}_1 \equiv \{1,3\}$, and $\mathbf{f}_2 \equiv \{2,4\}$ we obtain all classical probabilities by identifying $i \to \lambda_i > 0$. The respective conditional probabilities are

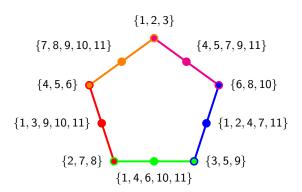
$$\begin{split} & [P(\mathcal{C}_{2}|\mathcal{C}_{1})] = [P(\{f_{1},f_{2},f_{3}\}|\{e_{1},e_{2},e_{3}\})] \\ & \equiv \begin{bmatrix} \frac{P(\{1\})}{P(\{1,2\})} & \frac{P(\{2\})}{P(\{1,2\})} & \frac{P(\emptyset)}{P(\{1,2\})} \\ \frac{P(\{3\})}{P(\{3,4\})} & \frac{P(\{4\})}{P(\{3,4\})} & \frac{P(\emptyset)}{P(\{3,4\})} \\ \frac{P(\emptyset)}{P(\{5\})} & \frac{P(\emptyset)}{P(\{5\})} & \frac{P(\{5\})}{P(\{5\})} \end{bmatrix} = \begin{bmatrix} \frac{\lambda_{1}}{\lambda_{1}+\lambda_{2}} & \frac{\lambda_{2}}{\lambda_{1}+\lambda_{2}} & 0 \\ \frac{\lambda_{3}}{\lambda_{3}+\lambda_{4}} & \frac{\lambda_{4}}{\lambda_{3}+\lambda_{4}} & 0 \\ 0 & 0 & 1 \end{bmatrix}, \end{split}$$
(3)

as well as

$$\begin{aligned} & [P(\mathcal{C}_{1}|\mathcal{C}_{2})] = [P(\{\mathsf{e}_{1},\mathsf{e}_{2},\mathsf{e}_{3}\}|\{\mathsf{f}_{1},\mathsf{f}_{2},\mathsf{f}_{3}\})] \\ \equiv \begin{bmatrix} \frac{P(\{1\})}{P(\{1,3\})} & \frac{P(\{3\})}{P(\{1,3\})} & \frac{P(\emptyset)}{P(\{1,3\})} \\ \frac{P(\{2\})}{P(\{2,4\})} & \frac{P(\{4\})}{P(\{2,4\})} & \frac{P(\emptyset)}{P(\{2,4\})} \\ \frac{P(\emptyset)}{P(\{5\})} & \frac{P(\emptyset)}{P(\{5\})} & \frac{P(\{5\})}{P(\{5\})} \end{bmatrix} = \begin{bmatrix} \frac{\lambda_{1}}{\lambda_{1}+\lambda_{3}} & \frac{\lambda_{3}}{\lambda_{1}+\lambda_{3}} & 0 \\ \frac{\lambda_{2}}{\lambda_{2}+\lambda_{4}} & \frac{\lambda_{4}+\lambda_{4}}{\lambda_{2}+\lambda_{4}} & 0 \\ 0 & 0 & 1 \end{bmatrix}. \end{aligned}$$
(4)

Examples: classical probabilities on House/Pentagon/Pentagram hyperdiagram

Hypergraph of the pentagon/pentagram/house logic with partition logic labelling (also available: vertex labelling by vectors aka faithful orthogonal representations).



Examples: classical probabilities on House/Pentagon/Pentagram hyperdiagram cntd.

With the identifications $e_1 \equiv \{1,2,3\}$, $e_2 \equiv \{4,5,7,9,11\}$, $e_3 \equiv \{6,8,10\}$, $f_1 \equiv \{2,7,8\}$, $f_2 \equiv \{1,3,9,10,11\}$, and $f_3 \equiv \{4,5,6\}$. The respective conditional probabilities are

$$\equiv \begin{bmatrix} \frac{P(\{2,7,8\} \cap \{1,2,3\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,10,11\} \cap \{1,2,3\})}{P(\{1,2,3\})} & \frac{P(\{4,5,6\} \cap \{1,2,3\})}{P(\{1,2,3\})} \\ \frac{P(\{2,7,8\} \cap \{4,5,7,9,11\})}{P(\{4,5,7,9,11\})} & \frac{P(\{1,3,9,10,11\} \cap \{4,5,7,9,11\})}{P(\{4,5,7,9,11\})} & \frac{P(\{4,5,6\} \cap \{1,2,3\})}{P(\{1,2,3\})} \\ \frac{P(\{2,7,8\} \cap \{6,8,10\})}{P(\{6,8,10\})} & \frac{P(\{1,3,9,10,11\} \cap \{6,8,10\})}{P(\{6,8,10\})} & \frac{P(\{4,5,6\} \cap \{6,5,7,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{2\})}{P(\{1,2,3\})} & \frac{P(\{1,3\})}{P(\{1,2,3\})} & \frac{P(\{0,1,3\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,10,11\} \cap \{6,8,10\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,10,11\} \cap \{6,8,10\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{2\})}{P(\{1,2,3\})} & \frac{P(\{1,3\})}{P(\{1,2,3\})} & \frac{P(\{0,1,3\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \\ \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{2\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \\ \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{2\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,3,9,11\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{1,3,9,11\})}{P(\{1,3,9,11\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,3,9,11\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} & \frac{P(\{1,3,9,11\})}{P(\{1,2,3\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,3,9,11\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{6,8,10\})} \end{bmatrix} \\ = \begin{bmatrix} \frac{P(\{1,3,9,11\})}{P(\{1,3,9,11\})} & \frac{P(\{1,3,9,11\})}{P(\{1,3,9,1\})} & \frac{P(\{1,3,9,11\})}{P(\{4,5,6\} \cap \{1,3,1\})} \\ \frac{P(\{1,3,9,11\})}{P(\{1,3,9,1\})} & \frac{P(\{1,3,1\})}{P(\{1,3,9,1\})} \\ \frac{P(\{1,3,9,1\})}{P(\{1,3,9,1\})} & \frac{P(\{1,3,1\})$$

 $[P(\mathcal{C}_2|\mathcal{C}_1)] = [P(\{f_1, f_2, f_3\} | \{e_1, e_2, e_3\})]$

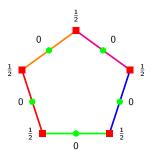
Examples: "exotic" probabilities on House/Pentagon/Pentagram hyperdiagram

Despite the aforementioned 11 two-valued states there exists another dispersionless state on cyclic pastings of an odd number of contexts; namely, a state being equal to $\frac{1}{2}$ on all intertwines/bi-connections; cf. Greechie DOI 10.1007/978-94-010-2274-3.

Wright DOI 10.1016/B978-0-12-473250-6.50015-7

This state and its associated probability distribution are neither realizable by quantum nor by classical probability distributions.

Hypergraph with overlaid "exotic" dispersionless state on the pentagon/pentagram/house logic:



Examples: "exotic" probabilities on House/Pentagon/Pentagram hyperdiagram

In this case the conditional probabilities of any two distinct contexts C_i and C_j , for $1 \le i, j \le 5$ are

$$[P(C_i|C_j)] \equiv \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}.$$
 (6)

Summary

Kolmogorov's axioms of probability theory are extended to conditional probabilities among distinct (and sometimes intertwining) contexts. Formally, this amounts to row stochastic matrices whose entries characterize the conditional probability to find some observable (postselection) in one context, given an observable (preselection) in another context. As the respective probabilities need not (but, depending on the physical/model realization, can) be of the Born rule type, this generalizes approaches to quantum probabilities by Aufféves and Grangier, which in turn are inspired by Gleason's theorem.

Thank you for your attention!